#### What is personalization worth for Netflix users? Evidence from a text-only and a mock VOD service choice experiment

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## Personalization vs. privacy: An inevitable trade-off?

(Awan and Krishnan 2006; Aguirre et al. 2016)

- Status Quo Netflix has gone from members choosing 2% of the recommended movies to 80% today (Biddle 2021).
- Personalization-privacy paradox:
  - Individuals continue to engage with personalized services despite expressing concerns.
  - Privacy is valued but willingly compromised for benefits such as time and effort savings, improved user experience, and enhanced content discovery.

We develop and use our recommender system because we believe that it is core to our business (Netflix 2021).

## Stated satisfaction

Netflix recommendations usually align with my taste.

I usually know what I want to watch on Netflix without relying on recommendations.



If you could decide, would you share this data with Netflix for personalized recommendations?

- NO:



# To what extent do you think the presented content is based on...?



#### Does Netflix follow you on socials?



• Information from other sources: We also obtain information from other sources. We protect this information according to the practices described in this Privacy Statement, plus any additional restrictions imposed by the source of the information. These sources vary over time, but could include:

• service providers that help us determine a location based on your IP address in order to customize our service and for other uses consistent with this Privacy Statement;

• security service providers that provide us with information to secure our systems, prevent fraud and help us protect the security of Netflix accounts;

• payment service providers that provide us with payment or balance information, or updates to that information, based on their relationship with you;

publicly-available sources such as publicly available posts on social media platforms and information available through public
 databases associating IP addresses with internet service providers (ISPs);

*Source*: https://help.netflix.com/legal/privacy

• Data-driven company? What beyond personalization?

# Do you know what would happen if you did not share this data?



### Data collection: choice experiments (2023)

- Choice Experiment 1
  - Data for money?
- Choice Experiment 2
  - Data for convenience?
- Two presentation formats:
  - Text-only experiment;
  - Mock App experiment

- Screening:
  - Netflix user,
  - At least one family member contributes to the subscription.
- Professional polling agency, June 2023, CAWI;
- N=50 (pilot study), N=1003 (text-only DCE), N=216 (mock DCE)
- Representative (gender and age);

# Credibility of hypothetical scenarios? New EU regulation to be introduced

- 1. Consent needed for the use of data for internal purposes,
- 2. Consent needed for the use of data for internal purposes, and to provide the service.

#### Increased credibility:

- Subscription plans in other countries,
- Ongoing regulatory debate,
- Payment vehicle to reduce protest votes;



#### Experiment 1: Data for money



#### Experiment 1: Data for time/convenience



## Results: Data for

#### money

- MXL, WTP Space, 1000 Sobol draws;
- 1 PLN = 0.25 USD;

- Substantial preference heterogeneity;
- MU>0 from "internal sharing"
- Very sensitive to the use of data from external sources;
- Against ads, and particularily personalized ads;
- Significant effect of the (experiential) presentation format;

		Mea	n effe	ct St. deviation				
var.	dist.	coef.		st. err.	coef.		st. err.	
Ratings	n	6.448	***	1.423	12.379	***	1.65	
Browsing history	n	6.339	***	1.055	-13.371	***	1.129	
Gender and age	n	-4.102	***	0.932	-2.271	*	1.697	
Location	n	-4.579	***	0.893	-9.749	**	4.131	
Socials	n	-10.968	***	1.574	12.542	***	1.952	
Pers. Ads	n	-7.117	***	1.429	18.87	***	1.567	
Ads Time	n	-1.953	***	0.477	2.651	***	0.5	
Discount	n	0.082	***	0.01	-5.146	***	0.01	
Mock Netflix effect:								
var.		coef.		st. err.				
Pers. Ads		-5.146	***	3.187				
Ads Time		-0.949	*	0.949				

Note: \*\*\*, \*\* and \* indicate 1%, 5% and 10% significance levels, respectively. n and ln indicate a random parameter distribution (either normal or log-normal).

#### Table 1. Results of the MXL (WTP space, in PLN)

# Results: Data for convenience

- MXL, Preference Space, 500 Sobol draws;
- Time ->
  - "I am not an average user" effect"?

- MU>0 from protecting data -> consistent with the previous experiment;
- Substantial preference heterogeneity
  - SD close to mean effects !!!
- V. ambigous effect of the "Mock App" experience;

#### Classes? ->

**Table 1.** Results of the MXL (Preference space)

		Mea	n effe	et	St. deviation			
var.	dist.	coef.		st. err.	coef.		st. err.	
ASC (disabling personalization)	n	-2.053	***	0.217	4.566	***	0.303	
Gender and age	n	0.563	***	0.06	0.188	*	0.112	
Location	n	0.773	***	0.068	-0.41	**	0.176	
Socials	n	1.063	***	0.080	-0.924	***	1.952	
Time & Convenience	n	-0.053	***	0.014	0.423	***	0.03	
Mock Netflix effect:								
var.		coef.		st. err.				
ASC (disabling personalization)		0.882	*	0.611				
Time & Convenience		-0.082	**	0.046				

Note: \*\*\*, \*\* and \* indicate 1%, 5% and 10% significance levels, respectively. n and ln indicate a random parameter distribution (either normal or log-normal).

### Results: Data for convenience (3-LC)

• LC, Preference Space, 3 classes;

Class 1 – now is good Class 2 – let's improve Class 3 – you won't get my data!

Mean effects:		1 CLASS				2 CLASS			3 CLASS		
var.	dist.	coef.		st. err.	coef.		st. err.	coef.		st. err.	
ASC (disabling personalization)	n	-2.0	***	0.105	-1.434	***	0.426	2.517	***	0.178	
Gender and age	n	-0.781	***	0.07	1.01	***	0.214	-2.83	***	0.375	
Location	n	-1.053	***	0.072	0.996	***	0.233	-2.299	***	0.343	
Socials	n	-1.326	***	0.094	0.945	***	0.233	-3.001	***	0.442	
Time & Convenience	n	-0.025	***	0.009	0.175	***	0.03	0.028	***	0.015	
Delta					-0.942	***	0.088	-0.295	***	0.067	
Mean prob.		0.4685			0.	0.1826			0.3489		

 Table 1. Results of the 3-class LC model (Preference space)

Note: \*\*\*, \*\* and \* indicate 1%, 5% and 10% significance levels, respectively.

### Conclusions

- Discrepancy between user stated preferences and the perspective presented by Netflix executives.
- Data other than ratings and browsing history remains sensitive.
- Discounts to incentivize users to share more data.
- Users are much more likely to accept random ads than personalized ads.
- Stated preferences may hinge on *"*experienced disutility".
- Substantial preference heterogeneity regarding the personalization-privacy trade-off.
- Next steps one more wave with more intensified "in-app" experience.

### Thank you for your attention!

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